#### **CINECA SCAI SuperComputing Application & Innovation**

## EuroHPC declaration

Slides stolen from several people presentations



Sanzio Bassini – May 2018

#### May, 2018

## Why invest in HPC?

HPC is at the core of major advances and innovations in the digital age

#### Strategic value for science

#### **HPC enables breakthrough science**

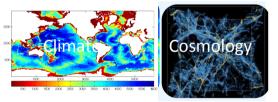
disease treatment; new therapies; brain; climate; chemistry; new materials; cosmology, astrophysics; high-energy physics; environment; transportation, earthquakes, etc.,

#### Strategic value for Industry

Market potential: new products, design and production cycles, decision processes, costs, resource efficiency, etc.

<u>National security and defense</u> Complex encryption technologies, terrorism, forensics cyber attacks, nuclear simulations





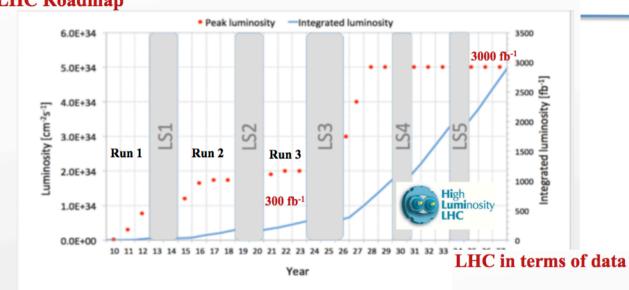


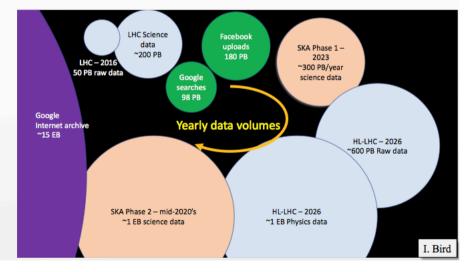






## High Energy physics

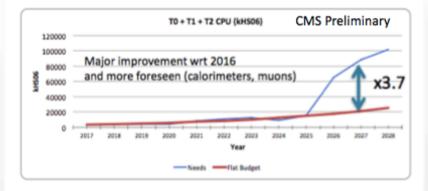




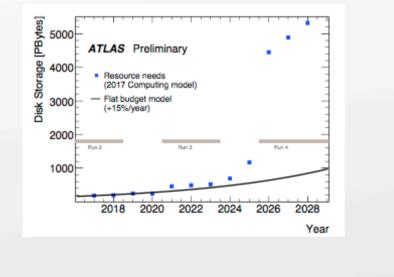
#### LHC Roadmap

#### **Resources Needs**

- In principle computing resources increase ~ linearly with data.
- Physicists already adopted several improvements, at the moment it is needed:
- factor ~ 4 of more CPU respect to the "20% per year growth"



 factor ~5 of more storage, cold storage (inactive data rarely used or accessed) concept already embedded



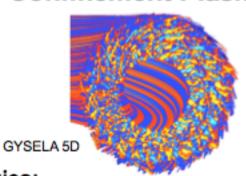
Donatella Lucchesi 4 May 2018 4

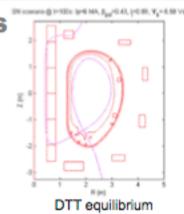
## **Nuclear Fusion Advanced Computing**

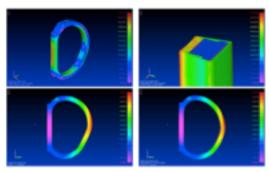
#### **Nuclear Fusion reactor with Magnetic Confinement Plasmas**

#### Plasma Physics:

- Turbolence (Gyro-kinetic codes), Edge
- MHD (Equilibrium, Transport, Instabilities)
- Heating, Fast particles





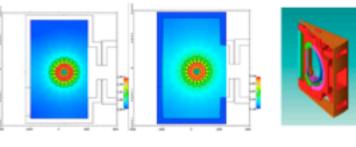


DTT: 3D stress analysis

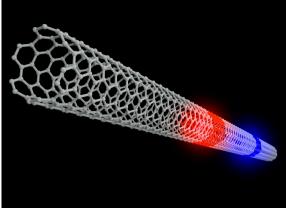


#### Reactor technologies:

- Neutron Transport: MCNP: Montecarlo N-Particle Transport
- Materials: DFT: Density Functional Theory for Radiation Damage
- Structural Analysis : FEM (Ansys, Comsol Multiphisics)



DTT: 3D MCNP analysis



### Carbon nanotubes as excitonic insulators

The **excitonic insulator** phase (EI), i.e. the instability of a zero gap semiconductor against the tendency of mutually attracting electrons and holes to form bound pairs, was speculated by W. Kohn in 1968 since then, the observation of the EI has remained elusive.

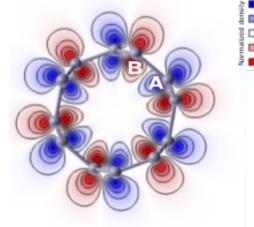
Here we have proved by means state of the art ab initio calculations

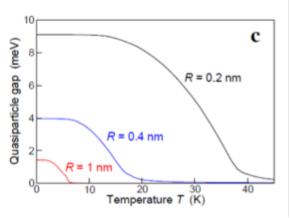
- through many-body perturbation theory as well as Quantum
- Monte Carlo that the excitonic insulator is realised in zero gap carbon nanotubes (CNT). The excitonic order modulates the charge between the two carbon sublattices opening an experimentally observable gap.

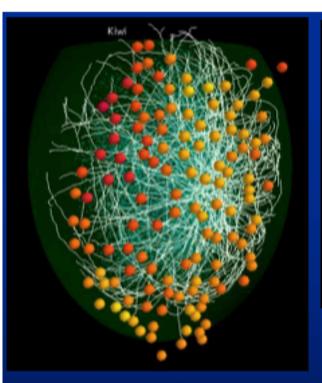
By means of mean field models using parameters provided from the ab initio calculations we observed that below a critical temperature the exciton phase is present in all the armchair family of CNT with an electronic gap scaling approximatively as the inverse of the tube radius.



Carbon nanotubes as excitonic insulators; D. Varsano, S. Sorella, D. Sangalli, M. Barborini, S. Corni, E. Molinari and M. Rontani; Nature Communications 8, 1461 (2017)











635 mitral cells 100K granule cells 7·10<sup>5</sup> synapses

(1/20 of the real system area 32,000,000 nonlinear ODEs)

Table 2 Model parameters and execution times for a typical simulation.

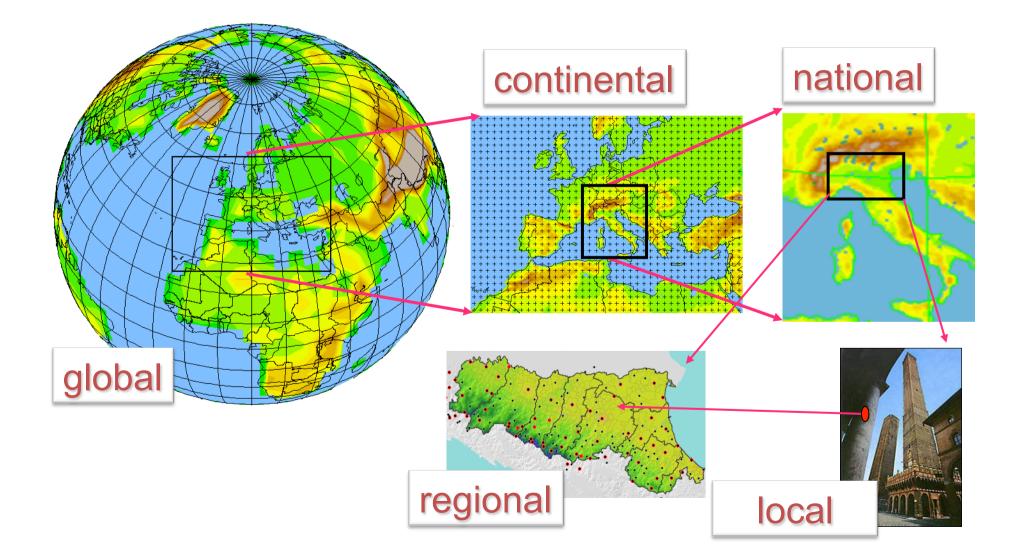
	Seg (min-max)	States (min-max) (v, ohan	Syn (min-max)				
MC ( $n = 635$ )	380,748 (189-1433)	5,259,705 (250)	707218 (300-2799)				
GC(n = 69013)	4,344,724 (33-257)	26,092,317 (2	707216 (1-62)				
Total	4,725,472	32,152,0	32,152,052				
	Computation time	Comm. time (spike exchange)	Comm. time (multisplit)	Total run time (2048 procs)			
Average (sec)	27149.35	68.53	555.94	32,552.86			
Max (sec)	27756.25	813.44	1453.96				

Currently installed on CINECA Marconi

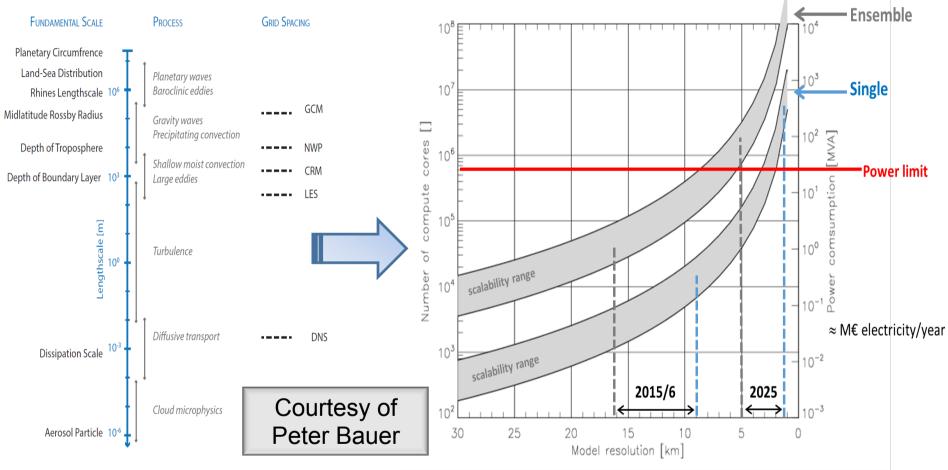
JSC JURECA, JUQUEEN

Typical 40 sec of sim. on 2048 processors, fully integrated NEURON+python implementation, 750.106 spikes: 9 hours, 10 Gb output, 99% eff.

Numerical Models to simulate climate change in different scenarious: Regionalization by dynamical or statistical downscaling



## Model resolution – computing Need of HPC



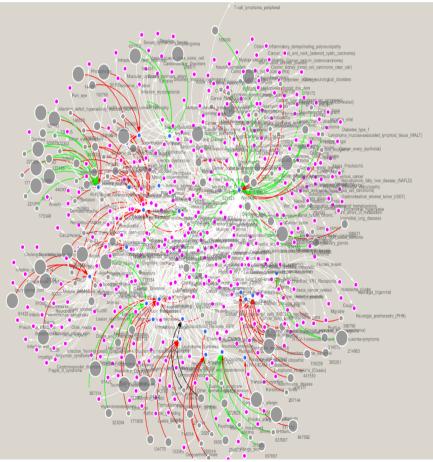
<sup>(</sup>Bauer et al. 2015)

→ Science community agrees that at very high resolution models will make qualitative jump in accuracy, but this comes at a very high computing & data cost, and that we estimate to be a factor of 1000 short with current system





Approaching disease-specific poly-pharmacology by connecting drugs with biological targets and diseases



Genome-wide molecular docking simulations will predict

- Drug Efficacy
- Drug Safety
- Averse Effects
- Novel Use of Known Drugs



# LiGen<sup>™</sup> Push to the limit software optimization on standard hardware



- Develop energy and resource efficient algorithms
- Use self-functionalities to adapt and scale-out the application

#### As Is

- Ligand: 100 Million
- Targets: 1
- Docking time: 30 sec
- 2048 CPU cores
- UNICODE / Flat file
- Wall Time: 407 hours (17 Days)
- Total Cost\*: 83K

- To Be
- Ligand: 1 Billion
- Targets: 10
- Docking time: 3 sec
- 1 Million CPU / GPU / MIC cores
- Binary / Database
- Wall Time: 8 hours
- Total Cost\*: 833K

Total Cost = 100 M Ligands X 1 Targets X 30/sec for each Dock X 0.1 Euro /hour

#### **Development Roadmap of LiGen™**

- Better Database Architectures
- Faster I/O
- Deeper Hardware / Software integration
- Ad Hoc Hardware
- New Evaluation Functions Based on Artificial Intelligence



ME

BROWSE CARS

## DATA CENTER 21 PetaByte

## BMW to make limited selfdriving available in 2021

What are you looking for?

In a presentation about its self-driving development, BMW said it would enable its cars to drive themselves on highways in 2021.

## Data 25.000.000 KM

Auto Tech

ROAD SHOW



Launch Your Own Cryptocurrency - Simple Token ICO Live Now

Cryptocurrency to power digital communities and build dynamic ecosystems sale.simpletoken.org

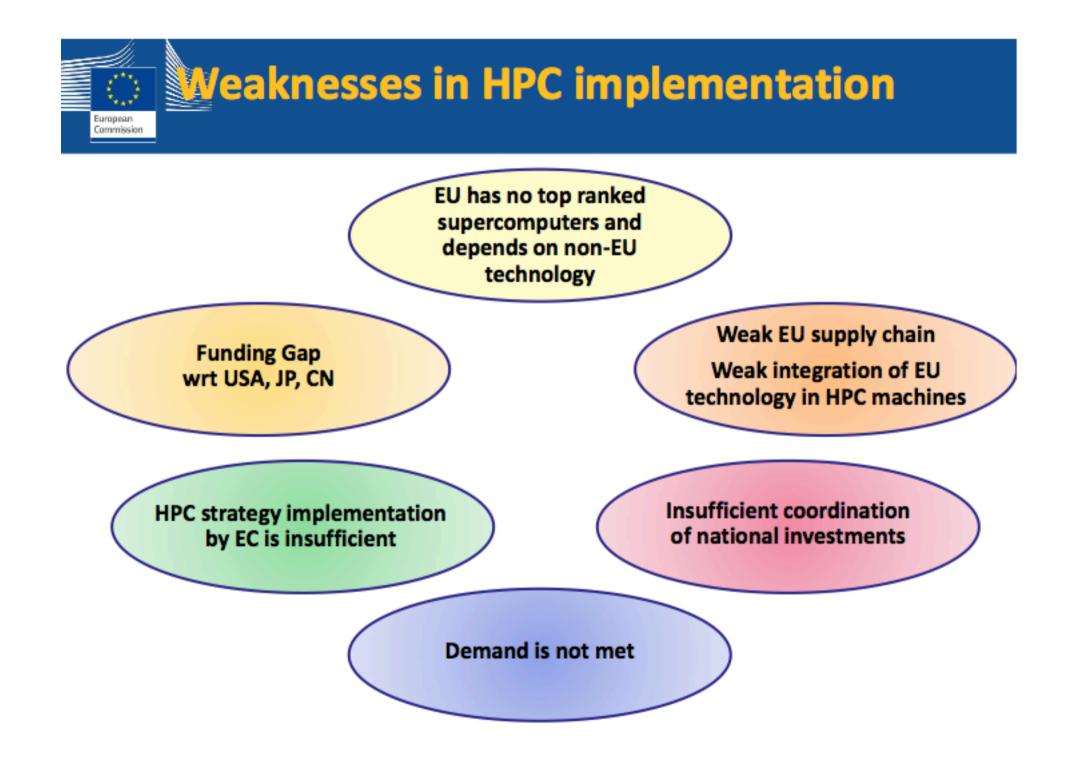
**Artificial Intelligence** 

SUBSCRIBE

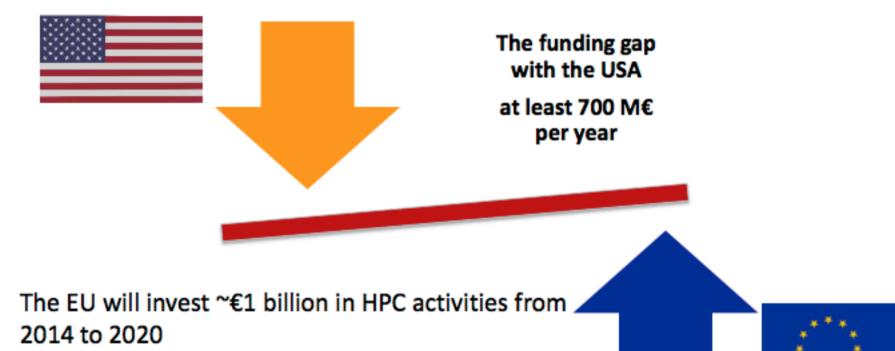
Given the sophistication of adaptive cruise control and lane-keeping assistance in recent **BMW** cars, such as the **M760**, you would think the company is well on its way to fully **self-driving cars**. During a presentation on its self-driving timeline, however, Klaus Büttner, BMW's vice president of autonomous driving, emphasized a more cautious approach.

The next step for BMW will come in

>







- US R&D investments: \$1 to \$2 billion per year
- China: over \$1 billion per year
- Japan: \$1.38 billion for 1 exascale system

No MS has the means to develop the necessary full HPC ecosystem on its own in competitive timeframes





## First estimation: €4.7 – 5.2 billion

COM(2016) 178 of 19/4/2016

2 pre-exascale and 2 exascale machines, data and interconnection

Technology development (processor, system, SW)

Applications + wider access to HPC facilities for SMEs

■ Until end 2020: €1B EC (H2020, CEF) + €1 B public & private sector

■ 2021 – 2026+: TBD

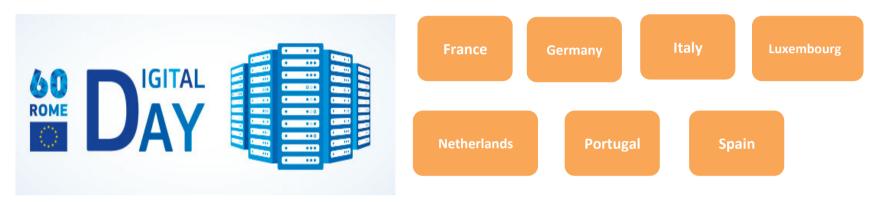
Overall envelope: ~ €8 – €9 billion

## **EuroHPC Declaration – Participating States**



Declaration signed in Rome 23/03/2017 by:

For Italy signed the Ministries of Education University and Research and of Economic Development



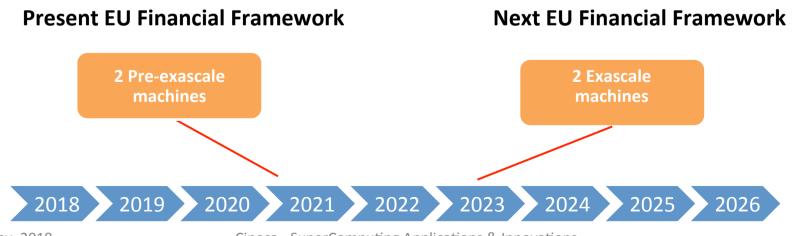
#### 8 more countries signed the Declaration:

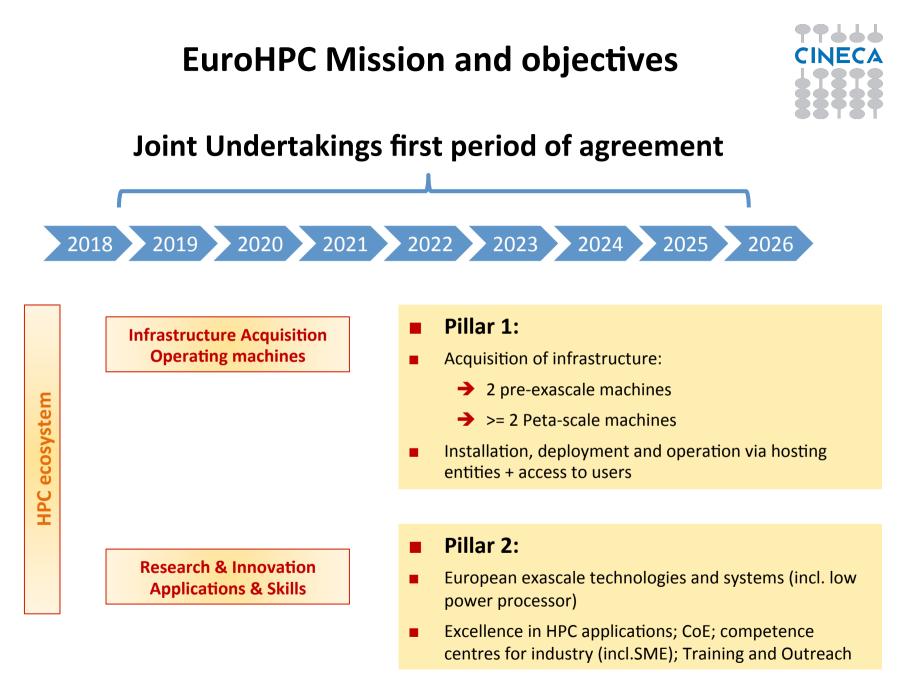


## **EuroHPC Mission and objectives**



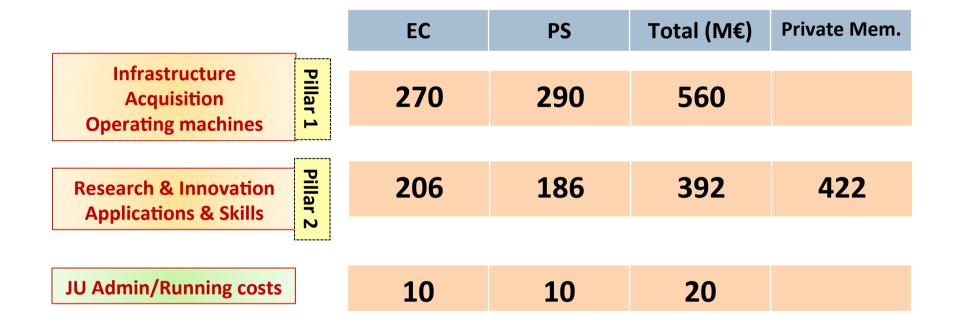
- to provide scientists, industry and the public sector from the Union with latest HPC and Data Infrastructure and support the development of its technologies and its applications across a wide range of fields.
- to provide a framework for acquisition of an integrated world-class preexascale supercomputing and data infrastructure in the Union;
- to provide Union level coordination and adequate financial resources to support the development and acquisition of such infrastructure, which will be accessible to users from the public and private sector primarily for research and innovation purposes;





### **Activites and Funding**



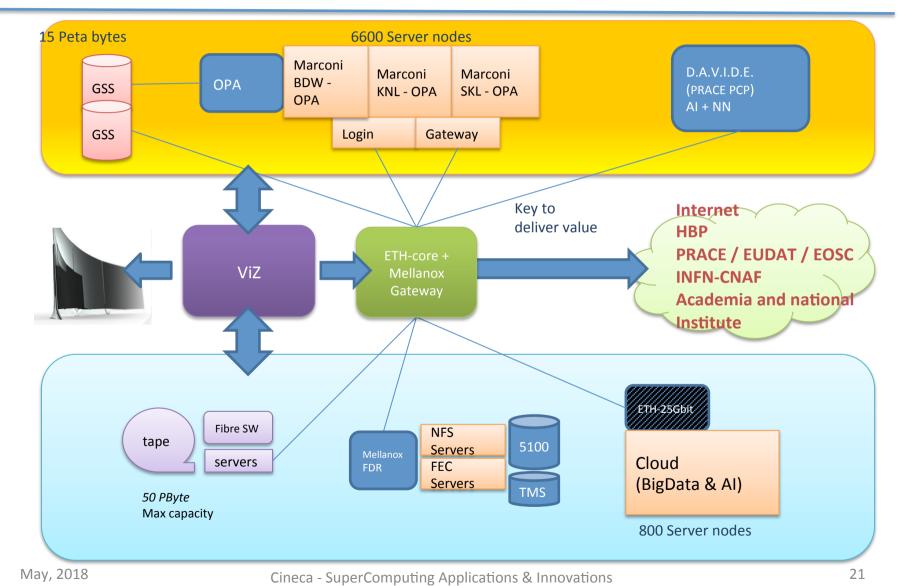




## National priorities for the JU workprograms and first estimates of national contribution

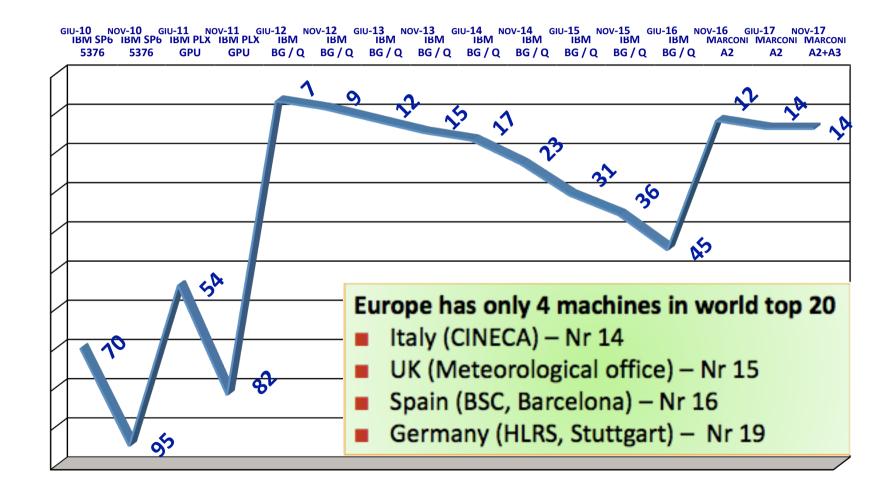


## HPC infrastructure design point



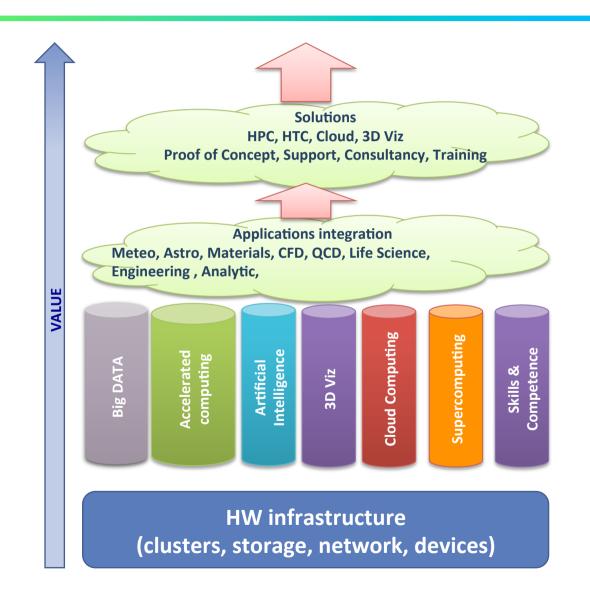


## Top500 (November 2017)

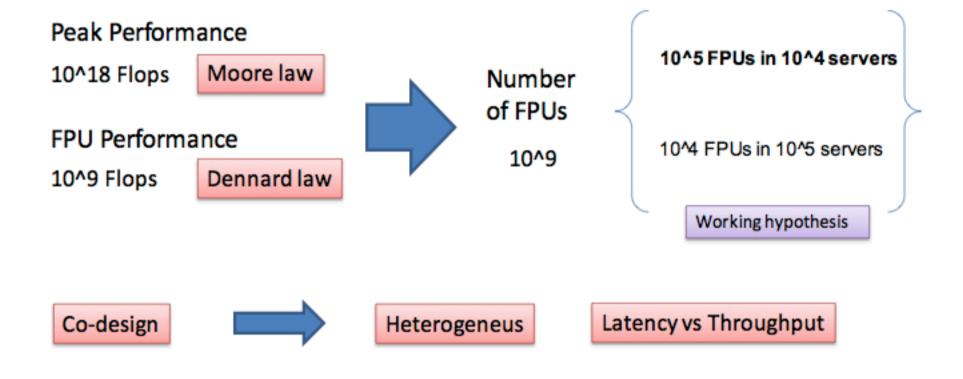


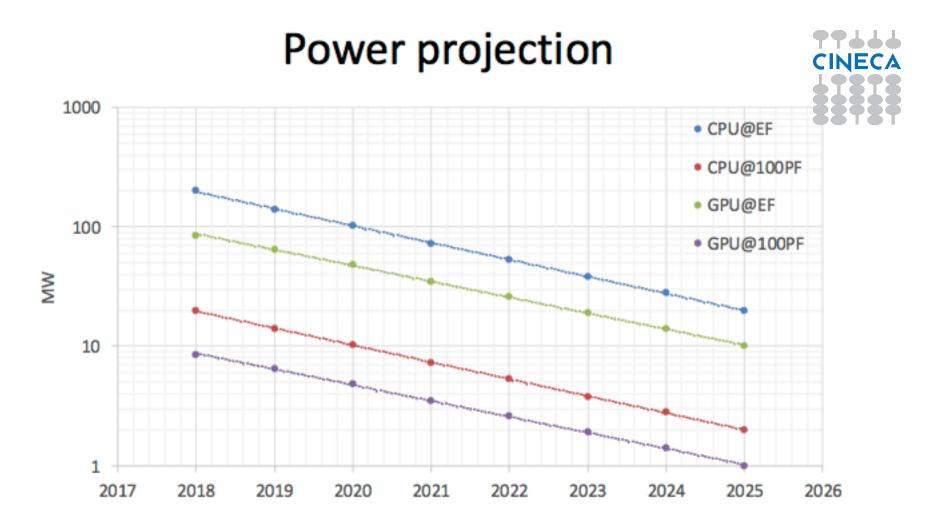
## HPC and Verticals





# Roadmap toward exascale



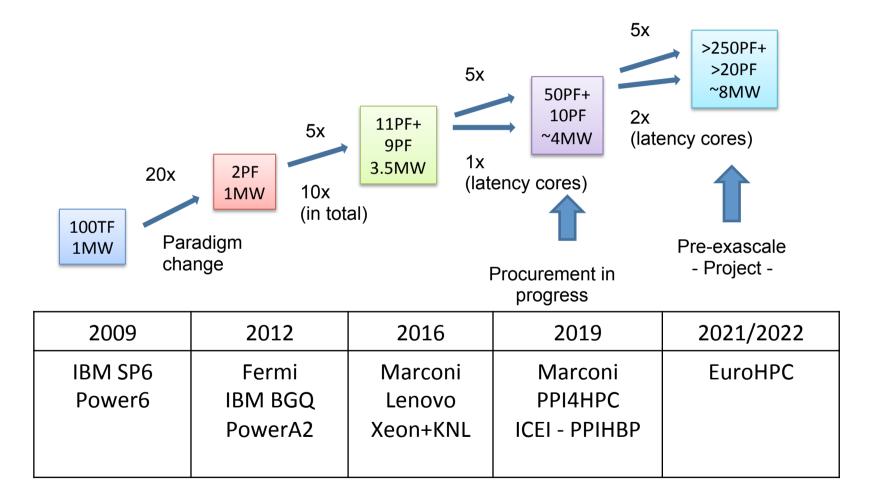


### Peek Perf (DP) @ 10MW

	2018	2019	2020	2021	2022	2023	2024	2025	2026
CPU	50PF	70PF	100PF	140PF	200PF	250PF	330PF	500PF	750PF
GPU	125PF	166PF	200PF	300PF	385PF	525PF	715PF	1EF	1.3EF

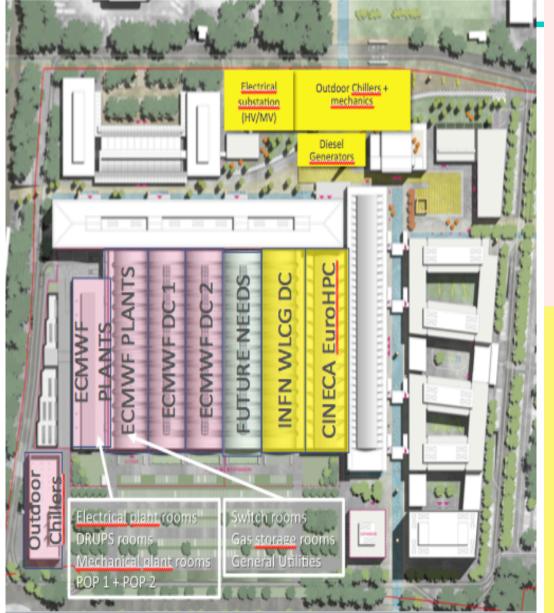






## **Towards Italian HE candidature**





#### **ECMWF DC main characteristics**

- 2 power line up to 10 MW (one bck up of the other)
- Expansion to 20 MW
- Photovoltaic cells on the roofs (500 MWh/ year)
- Redundancy N+1 (mechanics and electrical)
- 5 x 2 MW DRUPS
- Cooling
  - 4 dry coolers (1850 kW each)
  - 4 groundwater welles
  - 5 refrigerator units (1400 kW each)
- Peak PUE 1.35 / Maximum annualized PUE 1.18

#### **INFN – CINECA DC main characteristics**

- up to 20 MW (one bck up of the other)
- Possible use of Combined Heat and Power Fuel Cells Technology
- Redundancy strategy
- Cooling
  - dry coolers
  - groundwater welles
  - refrigerator units
- PUE < 1.2 1.3 / Max Annualized < 1.2 /
  - 1.17

# The CINECA-INFN datacentre timeline



	CINECA-CNAF datacentre timeline											
	2018				2019			2020				
	Q1	Q2	Q3	Q4	Q1	Q2	Q3	Q4	Q1	Q2	Q3	Q4
Agreement with region												
Tender for DC design												
DC design												
Tender for DC construction												
DC construction												
Derivation from 135KV												
External buildings												

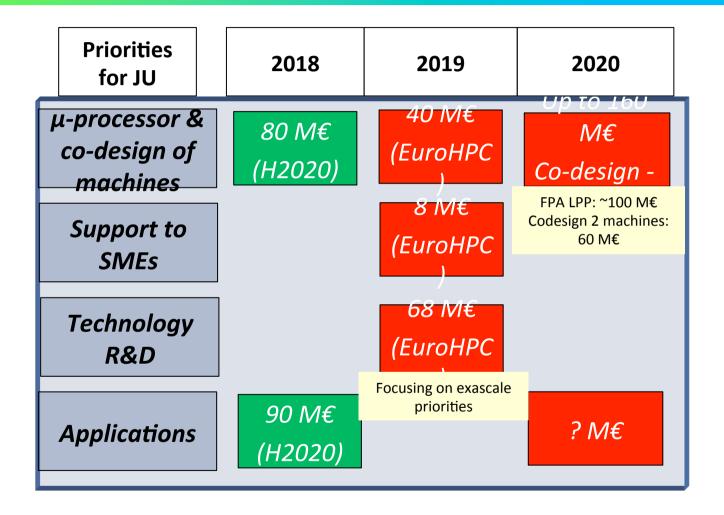
# Italian Estimates contribution for JU Pillar 1



- Solid national CINECA-INFN strategy and road-map for next 10 years
- Great DC location, with cross-fertilization opportunities
  - CINECA EuroHPC, INFN WLCG, ENEA DTT & Eurofusion, ECMWF, Italia Meteo, etc...
  - R&D labs of private companies
- 50% of TCO in kind contribution commitment for a pre-exascale system
- Commitment to represent an HE including others PS

# EU funding for R&I activities in 2019-2020





## Italian National Priority JU Pillar 2



Priorities for JU	Calls programmed 2019	Calls programmed 2020	Italian National priorities
μ-processor & co- design of machines	Framework Partnership Agreement in European Iow-power microprocessor technologies (Phase 2) <b>40 M€</b>	FETHPC-03-2020 : Co-design of extreme scale HPC systems and applications 160 M€	<ol> <li>Co-design for computing accelerators with standard interfaces: PCIe, OPENCAPI, ZENG</li> <li>Co-design of micro architecture of a computing efficient, exa-scalable direct network</li> <li>HW/SW support at micro architecture level to optimize the use of Machine Learning methods in scientific computing (either for HPC and HTC applications)</li> <li>HW/SW co-design with full applications</li> </ol>
Support to SMEs	INFRAINNOV-01-2019: Stimulate the innovation potential of SMEs 8 M€		<ol> <li>Public Private HPC Cloud Service</li> <li>HPC power efficiency with innovative solutions</li> <li>Academia and Research InstituteTechnology transfer initiatives to improve 3rd mission activities</li> <li>Industria 4.0 funded by Ministry of Economic Development</li> </ol>
Technology R&D	FETHPC-02-2019 : HPC - Extreme scale computing technologies, methods and algorithms for key applications and support to the HPC ecosystem <b>68 M€</b>		<ol> <li>Toward ecosystems for energy efficient - Power Management and Performance Monitoring</li> <li>Advanced interconnect topologies optimized for low latency and high throughput</li> <li>Advanced programming model and software tools for parallel ExaScalable scientific application</li> <li>Development of exascale oriented algorithms (eg mixed precision algorithms)</li> </ol>
Applications			<ol> <li>Increase funding on applications with respect to the co-design processes</li> <li>development of novel scientific and algorithm approaches able to take full advantage olarge scale HPC machines</li> <li>LQCD at extreme scale</li> <li>Bio-computing and Complex system simulation</li> <li>Neural network simulation and Neural network training for embedded solution</li> <li>Machine Learning methods applied to off-line/on-line particle tracking at HL-LHC experiments</li> </ol>

- 5) Machine Learning methods applied to off-line/on-line particle tracking at HL-LHC experiments
- 6) Quantum computing simulation and modeling